



CS 540 Introduction to Artificial Intelligence

Linear Algebra

University of Wisconsin–Madison

Fall 2025, Section 3

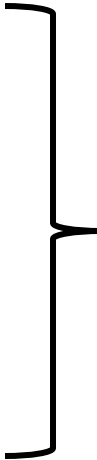
September 10, 2025

Announcements

- HW 1 to be released Friday (9/12)
- Optional Review Sessions
 - Held by TA Guy Zamir in Morgridge Hall 3610
 - This Thursday, 5:30-6:30 pm
 - From next week: Tuesdays at 5:30-6:30 pm

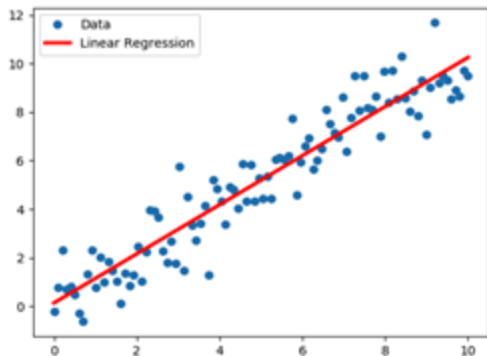
Announcements

- Class roadmap:

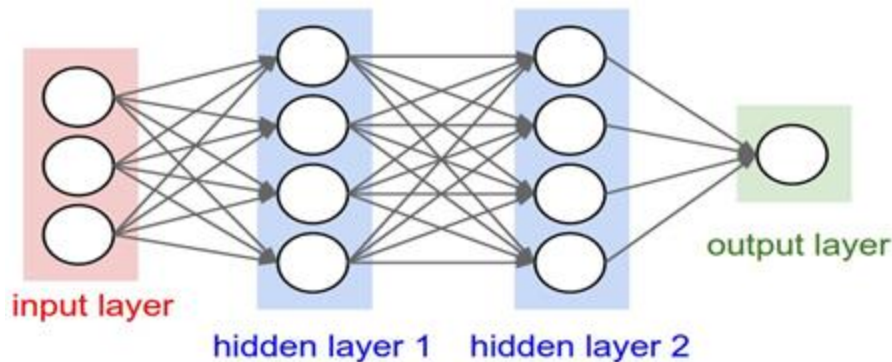
Probability	 Mostly Foundations
Statistics	
Linear Algebra	
Principal Component Analysis (PCA)	
Logic	
NLP	

Artificial Intelligence Runs on Linear Algebra

- Study of **linear** functions: simple, tractable
- In AI/ML: building blocks for all models
 - e.g., linear regression; part of neural networks



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Stanford CS231n

Artificial Intelligence Runs on Linear Algebra

Graphics Processing Units (GPUs) excel at linear algebra workloads



[nvidia.com](https://www.nvidia.com)



engineering.fb.com

Today: Linear Algebra Overview

- Basics
- Matrix Multiplication
- Advanced Topics

Mathematical tools for:

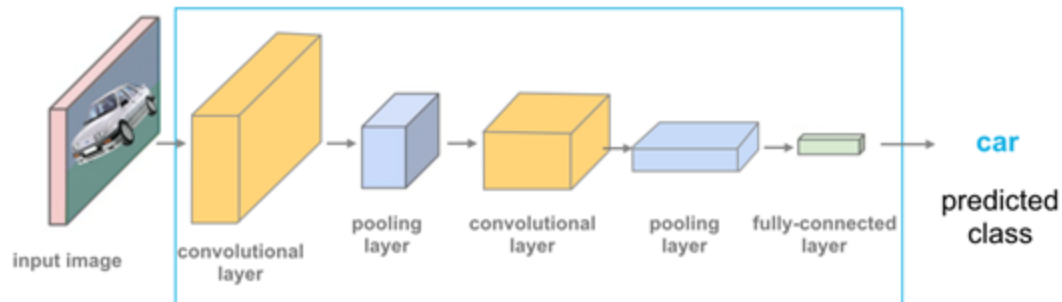
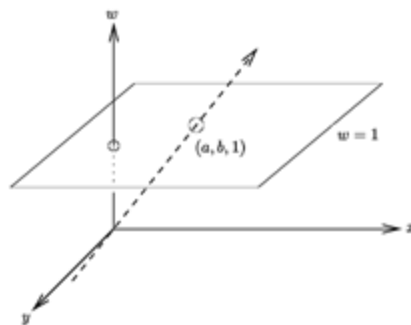
1. Organizing and manipulating high-dimensional data
2. Working with linear functions

Basics

Basics: Vectors

- Many interpretations
 - List of values (represents information)
 - **Point in a space**
- Dimension: number of values: $x \in \mathbb{R}^d$
- AI/ML: often use **very high dimensions**:
 - Ex: images!

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} \in \mathbb{R}^5$$



Basics: Matrices

- Many interpretations

- Array of values

- List of vectors

- Description of **linear transformation**

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \end{bmatrix}$$

Basics: Dimensions and Indexing

- Always: (# rows) x (# columns)
- Applies to vectors and scalars
- Indexing: same convention

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \end{bmatrix}$$

$$A \in \mathbb{R}^{4 \times 3}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$x \in \mathbb{R}^{2 \times 1} = \mathbb{R}^2$$

$$z = [z]$$

$$x \in \mathbb{R}^{1 \times 1} = \mathbb{R}$$

$$B = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

$$B_{21} = 3$$

Basics: Transposition

- Transpose: flip rows and columns

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \\ 7 & 8 \end{bmatrix} \quad A^T = \begin{bmatrix} 1 & 3 & 5 & 7 \\ 2 & 4 & 6 & 8 \end{bmatrix}$$

- Same for vectors
 - Standard vector is row
 - Transpose yields column

Basics: **Addition**

- Entry-by-entry

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} + \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix} = \begin{bmatrix} 1 + 5 & 2 + 6 \\ 3 + 7 & 4 + 8 \end{bmatrix} = \begin{bmatrix} 6 & 8 \\ 10 & 12 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} + \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 + 3 \\ 2 + 4 \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \end{bmatrix}$$

Matrix Multiplication

Gameplan for Matrix Multiplication

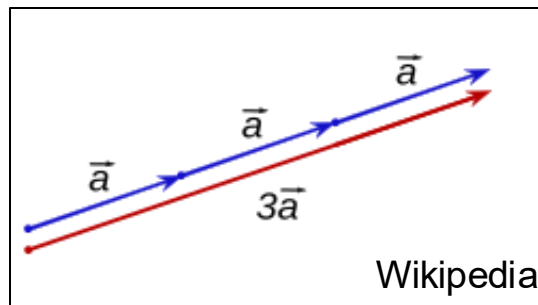
- Addition for matrices and vectors is simple.
- What about multiplication?
- Answer is counterintuitive!
- We will see:
 - Two simple cases
 - The general rule
 - More simple cases

Matrix Multiplication: Vector & Scalar

- How to multiply a vector and a scalar?

$$c\mathbf{x} = c \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} cx_1 \\ cx_2 \\ cx_3 \end{bmatrix}$$

- Uniform stretch/scaling on each component



Matrix Multiplication: Diagonal Matrix

- Recall: a matrix represents a linear transformation
- **Diagonal** matrices are simple transformations
- Stretch differently along different directions

$$D = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$Dx = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} d_{11}x_1 \\ d_{22}x_2 \\ d_{33}x_3 \end{bmatrix}$$

Scale different components separately!

Matrix Multiplication in General

- Rule: “middle” dimensions must match
 - To multiply $A \in \mathbb{R}^{n \times m}$ and $B \in \mathbb{R}^{p \times q}$, need $m = p$
 - Yields $C \in \mathbb{R}^{n \times q}$
- Computation for $C = AB$ is an *inner sum*

$$c_{ij} = \sum_{k=1}^p a_{ik} b_{kj}$$
$$\begin{bmatrix} c_{11} & c_{12} & c_{13} \\ c_{21} & c_{22} & c_{23} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \\ b_{31} & b_{32} & b_{33} \\ b_{41} & b_{42} & b_{43} \end{bmatrix}$$

Matrix Multiplication: Matrix-Vector

- Multiply

- $A \in \mathbb{R}^{n \times p}$

- $x \in \mathbb{R}^p = \mathbb{R}^{p \times 1}$

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_3 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

(4×1) (4×3) (3×1)

these must
match

result is in
 $\mathbb{R}^{4 \times 1}$

$$y_2 = a_{21}x_1 + a_{22}x_2 + a_{23}x_3$$

(4×1)

Matrix Multiplication: Outer Product

- Multiply vectors, get a matrix

$$xy^T = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \begin{bmatrix} y_1 & y_2 \end{bmatrix} = \begin{bmatrix} x_1 y_1 & x_1 y_2 \\ x_2 y_1 & x_2 y_2 \end{bmatrix}$$

$$x \in \mathbb{R}^2 = \mathbb{R}^{2 \times 1}$$

$$y \in \mathbb{R}^2 = \mathbb{R}^{2 \times 1}$$

$$y^T \in \mathbb{R}^{1 \times 2}$$

$$xy^T \in \mathbb{R}^{2 \times 2}$$

Matrix Multiplication: Inner Product

- Multiply vectors, get a **scalar**

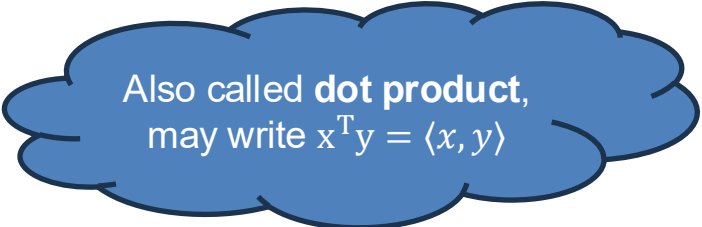
$$x^T y = [x_1 \quad x_2] \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = x_1 y_1 + x_2 y_2$$

$$x \in \mathbb{R}^2 = \mathbb{R}^{2 \times 1}$$

$$x^T \in \mathbb{R}^{1 \times 2}$$

$$y \in \mathbb{R}^2 = \mathbb{R}^{2 \times 1}$$

$$x^T y \in \mathbb{R}^{1 \times 1} = \mathbb{R}$$



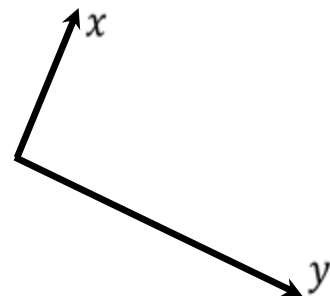
Also called **dot product**,
may write $x^T y = \langle x, y \rangle$

Norms and Orthogonality

- Vectors x and y are **orthogonal** if $\langle x, y \rangle = 0$
- Norms capture “length”

Euclidean norm

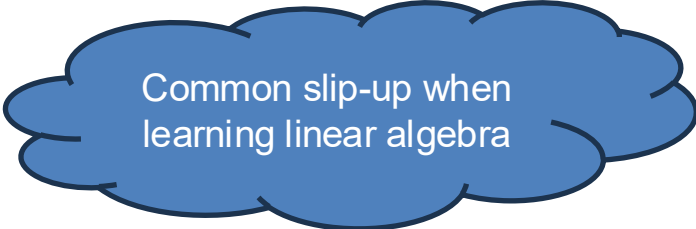
$$\|x\|_2 = \sqrt{\langle x, x \rangle} = \sqrt{\sum_{i=1}^d x_i^2}$$



- A set of vectors $\{x_1, x_2, \dots, x_n\}$ is **orthonormal** if:
 - For all pairs x_i, x_j , we have $\langle x_i, x_j \rangle = 0$
 - For all x_i , we have $\|x_i\|_2 = 1$

Matrices May Not Commute

- In general, $AB \neq BA$



Common slip-up when
learning linear algebra

Identity Matrix

- Like “1”
- Multiplying by it gets back the same matrix or vector
- Rows & columns are the “**standard basis vectors**” e_i

$$I = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \dots & \vdots \\ 0 & 0 & \dots & 1 \end{bmatrix}$$

$\downarrow \quad \downarrow \quad \quad \downarrow$
 $e_1 \quad e_2 \quad \quad e_n$

Matrix Inverse

- If there is a B such that $AB = BA = I$
 - Then A is invertible/nonsingular, B is its **inverse**
 - Some matrices are **not** invertible!
- Notation: A^{-1}

$$\begin{bmatrix} 1 & 1 \\ 2 & 3 \end{bmatrix} \times \begin{bmatrix} 3 & -1 \\ -2 & 1 \end{bmatrix} = I$$

Matrix & Vector Operations

Ex: feedforward neural networks. Input x .

- Output of layer k is

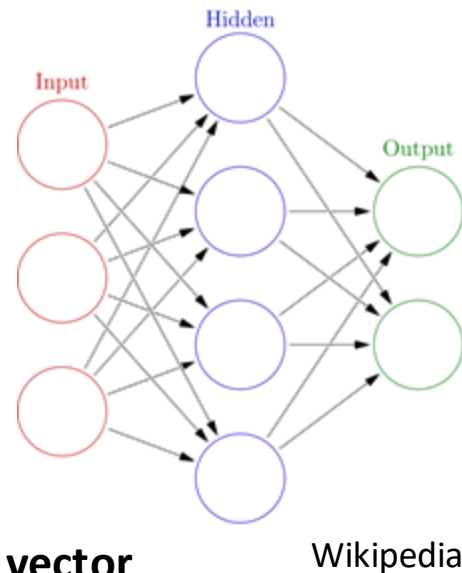
$$f^{(k)}(x) = \sigma(W_k^T f^{(k-1)}(x))$$

nonlinearity

Output of layer k-1: **vector**

Output of layer k: vector

Weight **matrix** for layer k:
Note: linear transformation!



Eigenvectors and Eigenvalues

Recall: Diagonal Matrix Multiplication

$$Dx = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} d_{11}x_1 \\ d_{22}x_2 \\ d_{33}x_3 \end{bmatrix}$$

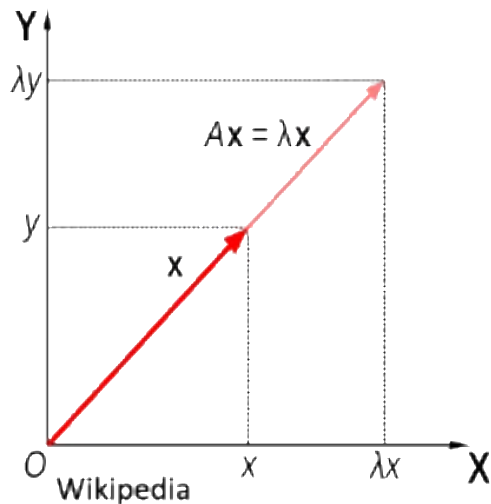
D stretches x entry-by-entry
Different components scale differently

Eigenvectors and Eigenvalues

- Sometimes, a matrix scales a vector uniformly
- Vector $x \neq 0$ is an **eigenvector** of A if there is a scalar λ such that

$$Ax = \lambda x$$

- λ is an **eigenvalue**
- A may have many pairs $\{(\lambda_1, x_1), \dots, (\lambda_d, x_d)\}$



Eigenvalues for Identity Matrix

- Definition: (λ, x) are an eigenvalue/eigenvector pair for A if $Ax = \lambda x$
- Identity matrix: for all x , $Ix = x$
- Thus, all vectors are eigenvectors for I ! (With $\lambda = 1$)

Eigenvalues for Diagonal Matrices

- Q: Which vectors satisfy $Dx = \lambda x$ for diagonal D ?

$$D = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix} \quad Dx = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} d_{11}x_1 \\ d_{22}x_2 \\ d_{33}x_3 \end{bmatrix}$$

- A: Standard basis vectors!

$$De_1 = \begin{bmatrix} d_{11} & 0 & 0 \\ 0 & d_{22} & 0 \\ 0 & 0 & d_{33} \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} d_{11} \\ 0 \\ 0 \end{bmatrix} = d_{11}e_1$$

Diagonals in Disguise

- Many matrices are **diagonalizable**
- We can find diagonal matrix D and orthonormal matrix Q such that

$$A = QDQ^{-1}$$

Q is orthonormal

Contain
eigenvectors of A

D is diagonal

Entries are
eigenvalues of A

Diagonals in Disguise

- Many matrices are **diagonalizable**
- We can find diagonal matrix D and orthonormal matrix Q such that

$$A = QDQ^{-1}$$

Orthonormal matrices:

1. Preserve length
2. Rotate & reflect

Diagonal matrices:

- Scale entry-by-entry

Further Reading

- Broad Reading: Gilbert Strang's notes & books
 - <https://math.mit.edu/~gs/LectureNotes/>
 - Introduction to Linear Algebra;
<https://math.mit.edu/~gs/linearalgebra/ila6/indexila6.html>
- Reading Ahead: math behind PCA
 - <https://web.stanford.edu/class/cs168/l/l7.pdf>
 - <https://web.stanford.edu/class/cs168/l/l8.pdf>

Linear Algebra

Basics

- Vector, Matrix
- Indexing
- Addition
- Transpose

Eigenvalues and Eigenvectors

- Satisfy $Ax = \lambda x$
- Diagonalize: $A = QDQ^{-1}$

Matrix Multiplication

- Dimensions must match
- Rule for computing
- Some special cases
- Not commutative!
- Inverse matrix
- Identity matrix
- Norm
- Orthogonal, Orthonormal

Mathematical tools for:

1. Organizing and manipulating high-dimensional data
2. Working with linear functions

Break & Quiz

- **Q 1.1:** What is $\begin{bmatrix} 1 & 2 \\ 3 & 1 \\ 1 & 1 \end{bmatrix} \times \begin{bmatrix} 0 \\ 1 \end{bmatrix}$?
- A. $[-1 \ 1 \ 1]^T$
- B. $[2 \ 1 \ 1]^T$
- C. $[1 \ 3 \ 1]^T$
- D. $[1.5 \ 2 \ 1]^T$

Break & Quiz

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- A. $[-1 \ 1 \ 1]^T$
- **B. $[2 \ 1 \ 1]^T$**
- C. $[1 \ 3 \ 1]^T$
- D. $[1.5 \ 2 \ 1]^T$

Check dimensions: answer must be 3 x 1 matrix (i.e., column vector).

$$\begin{bmatrix} 1 & 2 \\ 3 & 1 \\ 1 & 1 \end{bmatrix} \times \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 * 1 + 1 * 2 \\ 0 * 3 + 1 * 1 \\ 0 * 1 + 1 * 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}$$

Break & Quiz

- **Q 1.2:** Given matrices $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{d \times m}$, $C \in \mathbb{R}^{p \times n}$
What are the dimensions of BAC^T
- A. $n \times p$
- B. $d \times p$
- C. $d \times n$
- D. Undefined

Break & Quiz

- **Q 1.2:** Given matrices $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{d \times m}$, $C \in \mathbb{R}^{p \times n}$
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- A. $n \times p$
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- C. $d \times n$
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Break & Quiz

- **Q 1.2:** Given matrices $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{d \times m}$, $C \in \mathbb{R}^{p \times n}$
What are the dimensions of BAC^T

- A. $n \times p$
- **B. $d \times p$**
- C. $d \times n$
- D. Undefined

To rule out (D), check that for each pair of adjacent matrices XY , the # of columns of X = # of rows of Y

Then, B has d rows so solution must have d rows. C^T has p columns so solution has p columns.

Break & Quiz

- **Q 1.3:** A and B are matrices, neither of which is the identity. Is $AB = BA$?
- A. Never
- B. Always
- C. Sometimes

Break & Quiz

- **Q 1.3:** A and B are matrices, neither of which is the identity. Is $AB = BA$?
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- B. Always
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Break & Quiz

- **Q 1.3:** A and B are matrices, neither of which is the identity. Is $AB = BA$?
- A. Never
- B. Always
- **C. Sometimes**

Matrix multiplication is not necessarily commutative.



Thanks!